

Dynamic Data-Driven Event Reconstruction for Atmospheric Releases

Project Description

This is a collaborative LDRD Exploratory Research project involving four directorates – Energy & Environment, Engineering, NAI and Computation. The project seeks to answer the following critical questions regarding atmospheric releases – *How much material was released? When? Where? and What are the potential consequences?* Inaccurate estimation of the source term can lead to gross errors, time delays during a crisis, and even fatalities. We are developing a capability that seamlessly integrates observational data streams with predictive models in order to provide the best possible estimates of unknown source term parameters, as well as optimal and timely situation analyses consistent with both models and data. Our approach utilizes Bayesian inference and stochastic sampling methods (Markov Chain and Sequential Monte Carlo) to reformulate the inverse problem into a solution based on efficient sampling of an ensemble of predictive simulations, guided by statistical comparisons with data.

Expected Results

We are developing a flexible and adaptable data-driven event-reconstruction capability for atmospheric releases that provides (1) quantitative probabilistic estimates of the principal source-term parameters (e.g., the time-varying release rate and location); (2) predictions of increasing fidelity as an event progresses and additional data become available; and (3) analysis tools for sensor network design and uncertainty studies. Our computational framework incorporates multiple stochastic algorithms, operates with a range and variety of atmospheric models, and runs on multiple computer platforms, from workstations to large-scale computing resources. Our final goal is a multi-resolution capability for both real-time operational response and high fidelity multi-scale applications.

Mission Relevance

This project directly contributes to the laboratory's homeland and national security mission by addressing a critical need for atmospheric release event reconstruction tools to support the rapidly growing number of operational detection, warning, and incident characterization systems being developed and deployed by the Department of Homeland Security (e.g., BioWatch, BWIC system, national BioSurveillance Initiative) and Department of Energy (e.g., Nuclear Incident Response Team assets and deployments). The event reconstruction and sensor siting tools developed by this project are targeted for integration into the next-generation National Atmospheric Release Advisory Center and DHS's new Inter-agency Modeling and Atmospheric Analysis Center, based at LLNL.

FY04 Accomplishments and Results

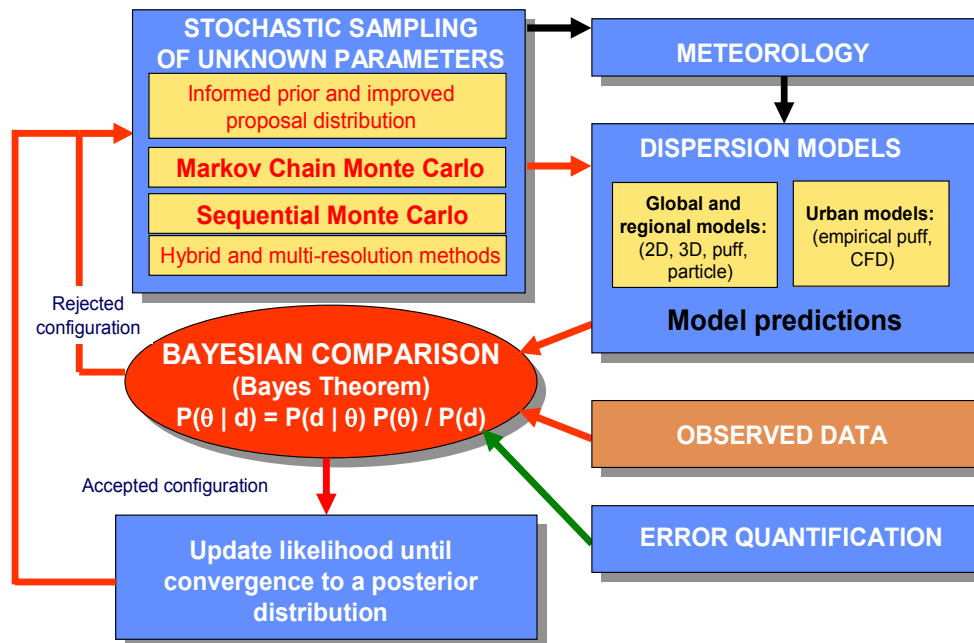
In FY04, we: (1) implemented a 2D prototype Bayesian Markov Chain Monte Carlo (MCMC) atmospheric event reconstruction capability and tested it against a variety of scenarios and synthetic data; (2) developed a MCMC capability using the NARAC 3D

Lagrangian particle dispersion code LODI and tested it against field experiment data; (3) developed a Sequential Monte Carlo (SMC) methodology for time-dependent problems and tested it using synthetic data; (4) investigated optimization methods for determining proposal distributions and improving convergence; (5) evaluated the relative performance of MCMC and SMC methods using synthetic data; (6) designed a MCMC and SMC computational framework on massively parallel platforms; and (7) demonstrated the effectiveness of the SMC methodology for detecting moving sources using synthetic data.

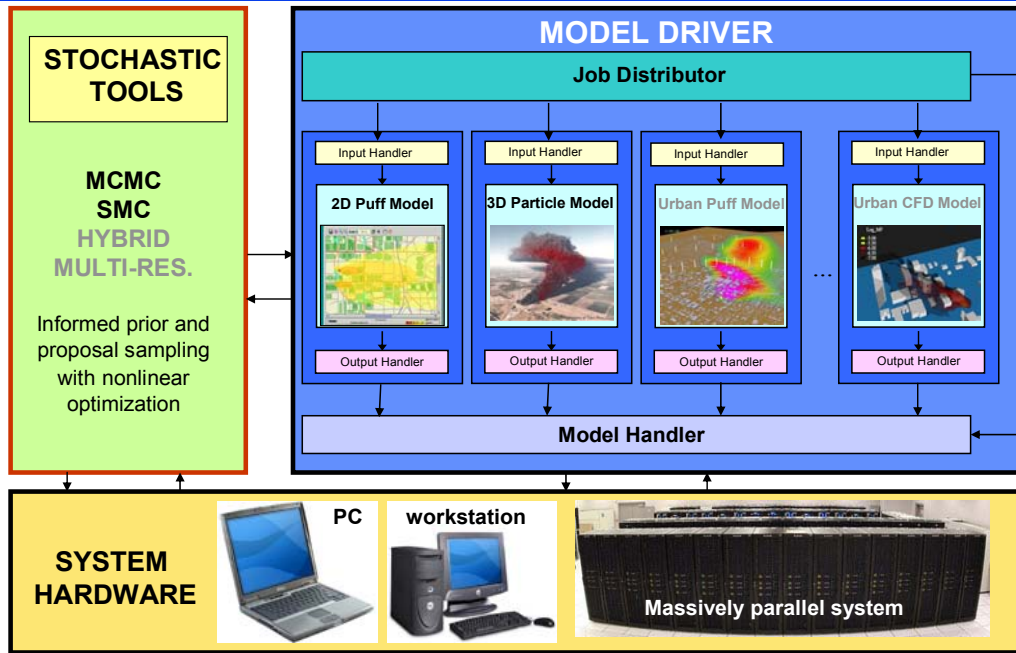
Proposed Work for FY05

In FY05 we will: (1) implement a SMC atmospheric event reconstruction capability based on NARAC operational models; (2) develop procedures and tools for input and model error quantification necessary for the event reconstruction framework; (3) extend the event reconstruction capability to handle complex terrain and an extended range of potential release types; (4) develop a multi-resolution capability for problems requiring multiple levels of models to accurately characterize the release event; (5) continue developing and testing efficient stochastic sampling and convergence algorithms; (6) explore methods for incorporating alternative input data types (e.g., remote sensing, qualitative, imagery); and (7) enhance performance on massively parallel platforms for efficient event reconstruction of complex atmospheric releases.

Predictive Models and Data are Coupled with Bayesian Inference and Stochastic Sampling



Computational Framework Will Support Multiple Stochastic Algorithms, Models, and Platforms



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This work was performed under the auspices of the U.S. Department of Energy by the University of California Lawrence Livermore National Laboratory under contract No. W-7405-ENG-48. This is LLNL Report UCRL-TR-210565.